

NPS55-79-07

NAVAL POSTGRADUATE SCHOOL

Monterey, California



STORAGE PROBLEMS WHEN DEMAND IS

"ALL OR NOTHING"

by

D. P. Gaver

and

P. A. Jacobs

March 1979

Approved for public release; distribution unlimited.

NAVAL POSTGRADUATE SCHOOL
MONTEREY, CALIFORNIA

Rear Admiral T. F. Dedman
Superintendent

J. R. Borsting
Provost

Reproduction of all or part of this report is authorized.

This report was prepared by:

UNCLASSIFIED

SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

REPORT DOCUMENTATION PAGE		READ INSTRUCTIONS BEFORE COMPLETING FORM
1. REPORT NUMBER NPS55-79-07	2. GOVT ACCESSION NO. AD-A070082	3. RECIPIENT'S CATALOG NUMBER
4. TITLE (and Subtitle) Storage Problems When Demand is "All or Nothing"		5. TYPE OF REPORT & PERIOD COVERED Technical
		6. PERFORMING ORG. REPORT NUMBER
7. AUTHOR(s) D. P. Gaver and P. A. Jacobs		8. CONTRACT OR GRANT NUMBER(s)
9. PERFORMING ORGANIZATION NAME AND ADDRESS Naval Postgraduate School Monterey, California		10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS
11. CONTROLLING OFFICE NAME AND ADDRESS Naval Postgraduate School Monterey, California		12. REPORT DATE March 1979
		13. NUMBER OF PAGES
14. MONITORING AGENCY NAME & ADDRESS (if different from Controlling Office)		15. SECURITY CLASS. (of this report) Unclassified
		15a. DECLASSIFICATION/DOWNGRADING SCHEDULE
16. DISTRIBUTION STATEMENT (of this Report) Approved for public release; distribution unlimited.		
17. DISTRIBUTION STATEMENT (of the abstract entered in Block 20, if different from Report)		
18. SUPPLEMENTARY NOTES		
19. KEY WORDS (Continue on reverse side if necessary and identify by block number) Inventorizes Bin-Packing Buffer Systems Communications Systems		
20. ABSTRACT (Continue on reverse side if necessary and identify by block number) An inventory of physical goods or storage space (in a communications system buffer, for instance) often experiences "all or nothing" demand: if a demand of random size D can be immediately and entirely filled from stock it is satisfied, but otherwise it vanishes. Probabilistic properties of the resulting inventory level are discussed analytically, both for the single buffer and for multiple buffer problems. Numerical results are presented.		

DD FORM 1 JAN 73 1473

EDITION OF 1 NOV 65 IS OBSOLETE
S/N 0102-014-6601

UNCLASSIFIED

SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

STORAGE PROBLEMS WHEN DEMAND IS "ALL OR NOTHING"

by

D. P. Gaver and P. A. Jacobs

Department of Operations Research
Naval Postgraduate School
Monterey, CA 93940

0. ABSTRACT

An inventory of physical goods or storage space (in a communications system buffer, for instance) often experiences "all or nothing" demand: if a demand of random size D can be immediately and entirely filled from stock it is satisfied, but otherwise it vanishes. Probabilistic properties of the resulting inventory level are discussed analytically, both for the single buffer and for multiple buffer problems. Numerical results are presented.

1. INTRODUCTION

The usual storage or inventory problems involve demands imagined to occur randomly, and to be capable of reducing any available stock to zero, or even beyond when backordering is permitted. Yet in many situations at least one component of total demand is "all or nothing;" that is, it reduces inventory only if it can be entirely satisfied by the inventory present, and otherwise seeks another supplier. Here are examples.

- (a) A manufacturer's warehouse is filled with a certain item at the beginning of the selling season; let I denote the initial inventory. Suppose that demands occur as follows: a message is sent requesting that D_1 items be shipped from inventory, but only if the entire order can be filled. That is, the demand is satisfied if $D_1 \leq I$, in which case inventory level is reduced to $I(1) = I - D_1$; while if $D_1 > I$ the inventory remains unchanged and $I(1) = I$. Allowing for no replenishment, the second demand, of size D_2 , interacts with inventory $I(1)$, so that it is filled if $D_2 \leq I(1)$, but is not placed if $D_2 > I(1)$. The process continues along these lines until the selling season is over and there are no more demands.
- (b) A buffer storage device used to contain messages prior to their batch transmission has capacity I . Messages of length $\{D_i, i = 1, 2, \dots\}$ approach the buffer successively, and are admitted on an "all or nothing" basis, just as was true of demands for physical inventory in (a) above. Once again rejection will occur, and more frequently to large demands (messages) than to short ones.
- (c) A system of many buffer storage devices is used to contain messages prior to their batch transmission. Each buffer has capacity I . Messages of length $\{D_i, i = 1, 2, \dots\}$ approach the device and are successively admitted to the

first buffer until there is a demand that exceeds its remaining capacity. The first buffer is left forever and the demand that exceeds the first buffer, plus successive demands, apply to the second buffer until one occurs that exceeds the remaining capacity. This demand then applies to the third buffer, and so on. As a result there will be some unused capacity in each buffer.

In Section 2 we will discuss some models for the situations in examples (a) and (b). We will compute such quantities as the distribution of the amount of inventory left at some time t and the distribution of the times of successive unsatisfied demands.

In Section 3 we will consider a model for example c. We will derive equations for the limiting distribution of used capacity of a buffer and the expected used capacity of a buffer. It seems to be difficult to obtain simple analytic solutions to these equations, but we will present certain illustrative numerical results.

2. THE ONE-BUFFER INVENTORY PROBLEM

Suppose that demands for available stock occur according to a compound Poisson process: if N_t is the number of demands that occur in $(0, t]$, then $\{N_t; t \geq 0\}$ is a stationary Poisson process with rate λ ; the sizes of successive demands $\{D_i\}$ are independent with common distribution F . Assume that there are no replishments of inventory. Let $\{I_t; t \geq 0\}$ denote the stochastic process describing available inventory at time t , and let $\{I(n); n = 0, 1, \dots\}$ be the stochastic process of available inventory following the n th demand. It is apparent from our assumptions that both $\{I_t\}$ and $\{I(n)\}$ are Markov processes.

2.1. Functional Equations for the Amount of Available Inventory

Let

$$(2.1) \quad \phi(s, t) = E[e^{-sI_t}]$$

be the Laplace transform of the available inventory at time t . Similarly, let

$$\psi(s, n) = E[e^{-sI(n)}] .$$

Properties of the available inventory can be studied in terms of ϕ and ψ ; we begin by deriving an equation for ϕ .

Observe that if one conditions on I_t , then in $(t, t+dt)$ either no change in inventory occurs, an event of probability $1 - \lambda F(I_t)dt + o(dt)$, or a depletion of amount x occurs with probability $\lambda dt F(dx)$, $x \leq I_t$. Thus

$$(2.2) \quad E[\exp(-sI_{t+dt}) | I_t] \\ = e^{-sI_t} [1 - \lambda F(I_t)dt] + \int_0^{I_t} \exp[-s(I_t - x)] \lambda dt F(dx) + o(dt) .$$

Now take expectations with respect to I_t :

$$(2.3) \quad \phi(s, t+dt) \\ = \phi(s, t) - \lambda \{ E[e^{-sI_t} F(I_t)] - E[\int_0^{I_t} \exp\{-s(I_t - x)\} F(dx)] \} dt + o(dt) .$$

After subtraction of $\phi(s, t)$ from both sides, division by dt , and allowing $dt \rightarrow 0$ we find that ϕ must satisfy the equation

$$(2.4) \quad \frac{\partial \phi}{\partial t} = \lambda E \left[e^{-sI_t} \int_0^{I_t} (e^{sx} - 1) F(dx) \right] .$$

An analogous argument shows that

$$(2.5) \quad \psi(s, n+1) = \psi(s, n) + E[e^{-sI(n)} \int_0^{I(n)} (e^{sx} - 1) F(dx)] .$$

Differentiation with respect to s at $s = 0$, or a direct conditional probability argument, now produce equations for $E[I_t]$ and $E[I(n)]$:

$$(2.6) \quad \frac{d}{dt} E[I_t] = -\lambda E\left[\int_0^{I_t} x F(dx)\right]$$

and

$$E[I(n+1)] = E[I(n)] - E\left[\int_0^{I(n)} x F(dx)\right].$$

In general no explicit solutions for the expected values are available, but a simple upper bound results from rewriting (2.6) as follows.

$$(2.7) \quad \begin{aligned} \frac{d}{dt} E[I_t] &= -\lambda E\left[I_t \int_0^{I_t} \frac{x}{I_t} F(dx)\right] \\ &\geq -\lambda E[I_t F(I_t)] \\ &\geq -\lambda F(I) E[I_t], \end{aligned}$$

from which one sees that

$$(2.8) \quad E[I_t] \geq I \exp[-\lambda F(I)t]$$

and similarly

$$E[I(n)] \geq I[1 - F(I)]^n,$$

so the expected available inventory declines by at most an exponential rate.

2.2. Explicit Solution When the Demand Distribution is Uniform

Although Equation (2.4) seems to be quite intractable for most demand distributions, it can be solved completely when F is uniform:

$$F(x) = \begin{cases} \frac{x}{c} & 0 \leq x \leq c, \\ 1 & c \leq x \end{cases}$$

and $c \geq I$. In this case (2.4) can be expressed as

$$\begin{aligned} (2.9) \quad \frac{\partial \phi}{\partial t} &= \lambda E \left[e^{-sI_t} \int_0^{I_t} (e^{sx} - 1) \frac{dx}{c} \right] \\ &= \lambda E \left[\frac{1 - e^{-sI_t}}{sc} - \frac{e^{-sI_t} I_t}{c} \right] \\ &= \frac{\lambda}{c} \left[\frac{1 - \phi}{s} \right] + \frac{\lambda}{c} \frac{\partial \phi}{\partial s}. \end{aligned}$$

In other words ϕ satisfies a first-order (quasi) linear partial differential equation with initial condition $\phi(s, 0) = e^{-sI}$. The usual procedure for solution, Sneddon (1957), requires solution of two ordinary differential equations selected from among

$$(2.10) \quad \frac{dt}{1} = \frac{-ds}{(\lambda/c)} = \frac{d\phi}{(\lambda/c) [(1-\phi)/s]};$$

we find from the first and last two that

$$(2.11) \quad s + \frac{\lambda}{c} t = c_1 , \quad \frac{1 - \phi}{s} = c_2 ;$$

so a general solution is given by

$$(2.12) \quad g\left(s + \frac{\lambda}{c} t, \frac{1 - \phi}{s}\right) = 0 ,$$

g being a function to be determined. Now the specified initial condition stipulated that at $t = 0$

$$(2.13) \quad \phi - e^{-sI} = 0 = g\left(s, \frac{1 - \phi}{s}\right) ,$$

so

$$(2.14) \quad \frac{1 - \phi}{s} - \frac{1 - e^{-sI}}{s} = 0$$

which specifies ϕ at $t = 0$. But for t positive we replace s , the first argument of g at $t = 0$ by $s + (\lambda/c)t$ to obtain the solution

$$(2.15) \quad \frac{1 - \phi(s, t)}{s} - \frac{1 - \exp[-(s + (\lambda/c)t)I]}{s + (\lambda/c)t} = 0 ,$$

which gives the desired transform. Passage to the limit as $s \rightarrow 0$ in (2.15) shows that

$$(2.16) \quad E[I_t] = \frac{1 - \exp[-(\lambda t/c)I]}{(\lambda/c)t} .$$

This formula can also be derived by first finding an expression for the kth moment of I_t , and then employing a Taylor series argument.

In order to invert the transform in (2.15) note that

$$(2.17) \quad \int_0^I e^{-sx} P\{I_t > x\} dx = \frac{1 - \phi(s,t)}{s} = \frac{1 - \exp[-(s + (\lambda t/c))I]}{s + (\lambda/c)t}$$

which is the transform of a truncated exponential distribution.

Thus by the unicity theorem for Laplace transforms

$$(2.18) \quad P\{I_t > x\} = \begin{cases} \exp[-(\lambda t/c)x] & 0 \leq x < I, \\ 0 & I \leq x . \end{cases}$$

Note that the distribution of I_t is absolutely continuous in the interval $(0, I)$ but that there is a jump at I corresponding to the occurrence of no demand less than or equal to I in $(0, t]$:

$$(2.19) \quad P\{I_t = I\} = \exp[-\lambda t(I/c)] .$$

2.3. The Expected Number of Satisfied Demands

Supposing that an initial inventory, or storage capacity, I prevails, it is of interest to compute the probability that a demand is satisfied, and the expected number of demands satisfied in an interval of length t . First notice that if a demand of size $D(t)$ appears at time t , at which moment I_t is available, then

$$P\{D(t) < I_t | I_t\} = F(I_t)$$

is the conditional probability that the demand is satisfied. When F is uniform, as is presently true, we may remove the condition to find that

$$P\{D(t) \leq I_t\} = E[F(I_t)] = E\left[\frac{I_t}{c}\right] = \frac{1 - \exp[-(\lambda t/c)I]}{\lambda t}.$$

If $S(t)$ is the number of demands satisfied during the time interval $(0, t]$, then since demands arrive according to a Poisson process with rate λ ,

$$\begin{aligned} (2.20) \quad E[S(t)] &= \lambda \int_0^t E[F(I_u)] du = \lambda \int_0^t \frac{1 - \exp[-(\lambda u/c)I]}{\lambda u} du \\ &= \gamma + \ln\left(\frac{\lambda t I}{c}\right) + E_1\left(\frac{\lambda t I}{c}\right) \end{aligned}$$

where $E_1(\cdot)$ is an exponential integral; Abramowitz and Stegun (1965), and $\gamma = 0.5772\dots$ is Euler's constant.

2.4. The Time of the First Unsatisfied Demand and the Amount of Unused Inventory at That Time

As before F is the common distribution function of the successive demands. Now let τ be the time of the first unsatisfied demand. Then

$$\begin{aligned} P\{\tau > t | N_t = n\} &= P\{D_1 \leq I, D_2 \leq I - D_1, \dots, D_n \leq I - D_1 - \dots - D_{n-1}\} \\ &= F^{(n)}(I) \end{aligned}$$

where $F^{(n)}$ denotes the n th convolution of F with itself.

Hence

$$(2.21) \quad P\{\tau > t\} = \sum_{n=0}^{\infty} e^{-\lambda t} \frac{(\lambda t)^n}{n!} F^{(n)}(I) .$$

Explicit expressions for the distribution of τ can be obtained in some cases. If F is uniform on $[0, c]$ with $c \geq I$, then

$$(2.22) \quad P\{\tau > t\} = e^{-\lambda t} I_0 \left[2 \left(\frac{\lambda t I}{c} \right)^{1/2} \right]$$

where $I_0(z)$ is a modified Bessel function of the first kind of the zeroth order. In this case

$$(2.23) \quad E[\tau] = \frac{1}{\lambda} \exp\{I/c\} = \frac{1}{\lambda} \exp\{I/2E[D]\} .$$

If F is exponential with mean $1/\mu$, then

$$(2.24) \quad P\{\tau > t\} = \sum_{n=0}^{\infty} e^{-\lambda t} \frac{(\lambda t)^n}{n!} \sum_{k=n}^{\infty} e^{-\lambda I} \frac{(\lambda I)^k}{k!}$$

and

$$(2.25) \quad E[\tau] = \frac{1}{\lambda} [1 + \mu I] = \frac{1}{\lambda} [1 + \frac{I}{E[D]}] .$$

Note that if I is small relative to $E[D]$, then the expected time to first unsatisfied demand when F is exponential will be greater than the expected time when F is uniform. However for I large relative to $E[D]$ the expected time for F exponential will be less than the expected time when F is uniform.

Let Y_n be the amount of inventory present at the time of the n^{th} unsatisfied demand. Then for $0 \leq a \leq I$

$$(2.26) \quad P\{Y_1 \geq I - a\} = \int_0^a R(dy) \bar{F}(I - y)$$

where

$$(2.27) \quad R(y) = \sum_{n=0}^{\infty} F^{(n)}(y)$$

and

$$(2.28) \quad \bar{F}(I - y) = 1 - F(I - y) .$$

Again explicit expressions for the distribution of Y_1 can be obtained for some distributions F . If F is uniform on $[0, c]$ for $c \geq I$, then

$$(2.29) \quad P\{Y_1 \geq I - a\} = 1 - \left(\frac{I - a}{c}\right) \exp\left\{\frac{1}{c} a\right\} .$$

If F is a truncated exponential

$$(2.30) \quad F(x) = \begin{cases} \frac{1 - e^{-\mu x}}{1 - e^{-\mu I}} & x \leq I , \\ 1 & x \geq I , \end{cases}$$

then

$$(2.31) \quad P\{Y_1 \geq I - a\} = 1 - [e^{-\mu a} - e^{-\mu I}] [1 - e^{-\mu I}]^{-1} \exp\{\mu a [1 - e^{-\mu I}]^{-1}\} .$$

If F has an exponential distribution with mean $1/\mu$, then

$$(2.32) \quad P\{Y_1 \geq I - a\} = e^{-\mu(I-a)} .$$

In this last case the distribution function of Y_n can be computed by induction quite easily and

$$(2.33) \quad P\{Y_n \geq I - a\} = e^{-n\mu(I-a)} .$$

Hence when F is exponential

$$(2.34) \quad E[Y_n] = \frac{1}{n\mu} [1 - e^{-n\mu I}] .$$

In principle similar results can be obtained for other distributions, but we have found no simple expressions.

2.5. Inventory Costs and Policies

There are at least three monetary quantities which affect the profitability of an inventory policy over a fixed interval of time $(0, t]$: the selling price, p ; the storage cost, a ; and the cost of lost demands, b . If the storage cost a is charged just on the basis of I (something like warehouse size) then the total expected profit in $(0, t]$ is

$$\begin{aligned} Z(I) &= p(I - E[I_t]) - aI - bS(t) \\ &= (p - a)I - p\left(\frac{\lambda}{c} t\right)^{-1} [1 - \exp[-(\lambda t/c)I]] \\ &\quad - b\{\gamma + \ln(\frac{\lambda t}{c} I) + E_1(\frac{\lambda t}{c} I)\} \end{aligned}$$

for the case of uniformly distributed demands; see ((2.16) and (2.20)). One can numerically find the maximum expected profit for this case; nothing explicit seems to be available.

3. THE MANY-BUFFER STORAGE PROBLEM

In this section we will study a model for the situation of example (c) in section 1. Messages are successively admitted to the n th buffer until there is a message length that exceeds the remaining capacity of the buffer. The total amount of this message is put in the $(n+1)$ st buffer and the n th buffer is left forever. Successive messages are then put in the $(n+1)$ st buffer until there is a message whose length exceeds the remaining capacity of the $(n+1)$ st buffer; this message is put in the $(n+2)$ nd buffer and so on.

Let I denote the common capacity of the buffers and D_i denote the length of message i . Assume $\{D_i\}$ is a sequence of independent identically distributed random variables with distribution F having a density function f such that $f(x) \geq d > 0$ for $x \in [0, I]$. Let $R(x) = \sum_{n=0}^{\infty} F^{(n)}(x)$ be the renewal function associated with F . If $F(I) < 1$, then we will assume that an incoming message to the currently used n th buffer of length greater than I is sent to the $(n+1)$ st buffer; when it cannot fit into the $(n+1)$ st buffer, then it is "banished," i.e. sent to some other set of buffers. The next message however will try to enter the $(n+1)$ st buffer. If this message has length greater than I it is banished and the following message will try to enter the $(n+1)$ st buffer; all messages of length exceeding I will be banished until one appears that is smaller than I and it will be the first entry in buffer $(n+1)$.

This model has been studied for demand distributions F with $F(I) = 1$ by Coffman et al. (1978). Their approach was to study the Markov process describing the total amount of inventory or space consumed in successive buffers or bins. Here we study the process $\{L_n\}$, where L_n is the size of the demand that first exceeds the remaining capacity of the n th buffer; $\{L_n; n = 1, 2, \dots\}$ is a Markov process. Let

$$K(x, [0, y]) = P\{L_{n+1} \leq y | L_n = x\}.$$

Note that

$$P\{L_1 \leq y\} = K(0, [0, y])$$

is the same as the sum of the forward and backward recurrence times at time I for a temporal renewal process with inter-renewal distribution F . Thus for $y \leq I$

$$(3.1) \quad H_1(y) \equiv P\{L_1 \leq y\} = \int_{I-y}^I R(dz) [F(y) - F(I-z)].$$

Note that for $y < I$

$$(3.2) \quad K(x, [0, y]) = \begin{cases} \int_{I-x-y}^{I-x} R(dz) [F(y) - F(I-x-z)] & \text{if } x < I-y; \\ \int_0^{I-x} R(dz) [F(y) - F(I-x-z)] & \text{if } I-y \leq x < I; \\ \int_{I-y}^I R(dz) [F(y) - F(I-z)] & \text{if } x > I. \end{cases}$$

Hence

$$(3.3) \quad K(x, dy) = \begin{cases} [R(I-x) - R(I-x-y)] F(dy) & \text{if } x < I-y, \\ R(I-x) F(dy) & \text{if } I > x > I-y, \\ R(y) F(dy) - \int_0^y R(dz) f(y-z) + R(dy)F(y) & \text{if } x = I-y, \\ [R(I) - R(I-y)] F(dy) & \text{if } x > I. \end{cases}$$

Note that for some $0 < a < b < I$, there exists a $\delta > 0$ such that for all x

$$K^2(x, dy) \geq \delta \quad \text{for } y \in [a, b]$$

where $K^2(x, dy) = \int_0^\infty K(x, dz) K(z, dy)$. Hence hypothesis D' on page 197 of Doob (1952) is satisfied. Thus, if

$$K^n(x, A) = P\{L_{1+n} \in A | L_1 = x\}$$

for all Borel subsets A , then

$$(3.4) \quad \lim_{n \rightarrow \infty} K^n(x, A) = H(A)$$

exists and further the convergence is geometric

$$|K^n(x, A) - H(A)| \leq \alpha \gamma^n$$

for some positive constants α and γ , $\gamma < 1$ for all A .

Now let

$$H_n(x) = P\{L_n \in [0, x] | L_0 = 0\}.$$

Then a renewal argument can be used to show that for $x \leq I$

$$(3.5) \quad H_{n+1}(x) = \int_{I-x}^I H_n * R(dy) [F(x) - F(I-y)] \\ + [1 - H_n(I)] \int_{I-x}^I R(dy) [F(x) - F(I-y)] .$$

Taking limits as $n \rightarrow \infty$ it is seen that the distribution $H(x)$ satisfies the following equation for $x \leq I$:

$$(3.6) \quad H(x) = \int_{I-x}^I H * R(dy) [F(x) - F(I-y)] \\ + [1 - H(I)] \int_{I-x}^I R(dy) [F(x) - F(I-y)] .$$

Equations (3.1) and (3.6) can be simplified for certain specific distributions F .

A. Exponential Demands.

For the exponential distribution with mean 1 and $x \leq I$ the equations are

$$(3.7) \quad H_1(x) = 1 - e^{-x} - xe^{-x}$$

and

$$(3.8) \quad H(x) = xe^{-x} H(I) + H_1(x) - e^{-x} \int_0^x H(I-x+u) du .$$

B. Uniform Demands

For the uniform distribution on $[0, c]$ with $c \geq I$ they simplify to

$$(3.9) \quad H_1(x) = \exp\left[\frac{1}{c}(I-x)\right] - \left(1 - \frac{x}{c}\right) \exp\left(\frac{1}{c}I\right)$$

and

$$(3.10) \quad H(x) = \frac{1}{c} \exp\left[\frac{1}{c}(I-x)\right] \int_0^{I-x} \exp\left(-\frac{1}{c}u\right) H(u) du \\ + \frac{x}{c} H(I) - \frac{1}{c} \exp\left(\frac{1}{c}I\right) \left(1 - \frac{x}{c}\right) \int_0^I \exp\left(-\frac{1}{c}u\right) H(u) du \\ + [1 - H(I)] H_1(x) ,$$

for $x \leq I$. Similar expressions hold for $x > I$, but they are unimportant in the present context.

Equations (3.6), (3.8) and (3.10) do not seem to yield explicit answers. As a result we have solved (3.8) and (3.10) numerically by iteration using the system of equations

$$(3.11) \quad H_{n+1}(x) = xe^{-x} H_n(I) + H_1(x) - e^{-x} \int_0^x H_n(I-x+u) du$$

with H_1 as in (3.7) and

$$(3.12) \quad H_{n+1}(x) = \frac{1}{c} \exp\left[\frac{1}{c}(I-x)\right] \int_0^{I-x} \exp\left(-\frac{1}{c}u\right) H_n(u) du \\ + \frac{x}{c} H_n(I) - \frac{1}{c} \exp\left(\frac{1}{c}I\right) \left(1 - \frac{x}{c}\right) \int_0^I \exp\left(-\frac{1}{c}u\right) H_n(u) du \\ + [1 - H_n(I)] H_1(x)$$

with H_1 as in (3.9). For the cases carried out the convergence

is rapid; after $n = 5$ iterations very little change is noted and convergence has occurred, for most practical purposes.

Next let Y_n be the amount of storage space used in the n th bin; the distribution of Y_n is denoted by $G_n(x)$, and

$$G(x) = \lim_{n \rightarrow \infty} P\{Y_n \leq x\} = \lim_{n \rightarrow \infty} G_n(x)$$

is the long-run distribution. By probabilistic arguments and (3.4)

$$(3.13) \quad G(x) = \int_0^x H * R(dy) \bar{F}(I-y) + [1-H(I)] \int_0^x R(dy) \bar{F}(I-y)$$

where $\bar{F}(I-y) = 1 - F(I-y)$ and the long run average expected capacity of a bin that is actually used is

$$A = \int_0^I x G(dx) .$$

For the case in which F is exponential with unit mean

$$(3.14) \quad A = I - [1-H(I)][1-e^{-I}] - e^{-I} \int_0^I e^x H(x) dx .$$

For the case in which F is uniform on $[0, c]$ with $c \geq I$

$$(3.15) \quad A = -2 \int_0^I H(u) du + \exp\left(\frac{1}{c} I\right) \int_0^I \exp\left(-\frac{1}{c} u\right) H(u) du \\ + H(I) [2I - c \exp\left(\frac{1}{c} I\right) + c] + [-I + c \exp\left(\frac{1}{c} I\right) - c] .$$

Numerical solutions were obtained for equations (3.14) and (3.15) by first computing the probabilities $H_n(x)$, $n = 1, 2, \dots, 10$ iteratively from (3.7) and (3.11) for the exponential demand case, and from (3.9) and (3.12) for the case of uniform demands. Our technique was simply to discretize x : $x_j = jh$, $h = I/N$, N being the number of x -values at which $H_n(x)$ is evaluated (values of N from 200-1200 were utilized in order to obtain two-significant digit accuracy). The integrals were then approximated by a summation, i.e. Simpson's rule. Having the values of $H_n(x_j)$ it is possible to calculate those of $H_{n+1}(x_j)$, and from these the values of $G_n(x)$ and the mean usage, $E[Y_n]$, may be calculated by numerical integration. In the case of exponential demand very simple upper and lower bounds were obtainable; such bounds were not tight enough to be useful for the uniform case.

The following table summarizes the numerical results. We have compared demand distributions that result, as nearly as possible, in the same probability that an initial demand on an empty bin will be rejected. We have tabulated the expected level to which the bin is filled. It is interesting that the limited bin occupancy is 0.75 when a uniform demand over the range of the bin size is experienced. This result has been obtained analytically by Coffman et al. (1978); in that paper simple and elegant analytical expressions for G and H also appear for this case. The considerable similarity of the numbers in the rows of the table is notable; apparently the long-run bin occupancy is only slightly larger than is that

of the first bin, and the occupancy experienced for uniform demand is only slightly larger than for exponential. Further investigations to examine the reasons for this insensitivity would seem to be of interest.

Expected Fraction of Bin Filled

$$(f_n = E[Y_n] \div I)$$

<u>Rejection Probability</u>	<u>Exponential Demand</u>		<u>Uniform Demand</u>	
$\bar{F}(I)$	f_1	f_∞	f_1	f_∞
0.00	-	-	0.76	0.75
0.05	0.74	0.75	0.74	0.74
0.10	0.69	0.70	0.72	0.72
0.15	0.65	0.66	0.68	0.69
0.20	0.60	0.62	0.64	0.66
0.25	0.56	0.58	0.60	0.62

Acknowledgments

The authors wish to acknowledge the research support of the National Science Foundation (NSF ENG 77-09020, ENG 79-01438 and MCS 77-07587) and the Office of Naval Research (Contract NR042-411. D. P. Gaver also wishes to acknowledge the hospitality of the Statistics Department, University of Dortmund (W. Germany), where he was a guest professor during the summer of 1977, and where part of this work was carried out.

REFERENCES

- Abramowitz, M. and Stegun I.A. (1965). Handbook of Mathematical Functions. National Bureau of Standards, AMS 55, Washington, D.C.
- Coffman, E.G., Jr., Hofri, M., and So, K. (1978). "A stochastic model of bin-packing." Technical Report TR-CSL-7811. Computer Systems Laboratory, University of California, Santa Barbara, Ca. 93106 (submitted to a technical journal).
- Cohen, J. W. (1977). Personal Communication.
- Doob, J. L. (1952). Stochastic Processes. John Wiley and Sons, N.Y.
- Feller, W. (1966). An Introduction to Probability Theory and Its Applications, Vol. II, John Wiley and Sons, N.Y.
- Sneddon, I. (1957). Elements of Partial Differential Equations, McGraw-Hill, N.Y.

DISTRIBUTION LIST

	No. of Copies
Defense Documentation Center Cameron Station Alexandria, VA 22314	2
Technical Information Division Naval Research Laboratory Washington, D.C. 20375	1
Library Code 0212 Naval Postgraduate School Monterey, CA 93940	2
Dean of Research, Code 012 Naval Postgraduate School Monterey, CA 93940	1
Statistics and Probability Program Office of Naval Research Code 436 Arlington, VA 22217	
Library Naval Ocean Systems Ctr San Diego, Ca. 92152	1
Naval Electronic Systems Command Navelex 320 National Center No. 1 Arlington, VA 20360	1
Director Naval Research Lab Attn: Library (ONRL) Code 2029 Washington, D.C. 20375	1
Office of Naval Research San Francisco Area Office 760 Market Street San Francisco, CA 94102	1
Prof. T. W. Anderson Department of Statistics Stanford University Stanford, CA 94305	1

	No. of Copies
Prof. F. J. Anscombe Department of Statistics Yale University New Haven, CT 06520	1
Prof. C. R. Baker Department of Statistics University of North Carolina Chapel Hill, N.C. 37512	1
Prof. R. E. Bechhofer Dept. of Operations Research Cornell University Ithaca, New York 14850	1
Prof. N. J. Bershad School of Engineering University of California, Irvine, CA 92664	1
P. J. Bickel Department of Statistics University of California Berkeley, CA 94720	1
Prof. H. W. Block Dept. of Mathematics University of Pittsburgh Pittsburgh, PA 15260	1
Prof. Joseph Blum Dept. of Math., Stat. and Computer Sci. The American University Washington, D.C. 20016	1
Prof. R. A. Bradley Dept. of Statistics Florida State University Tallahassee, FL 32306	1
Prof. R. E. Barlow Operations Research Ctr College of Engineering University of California Berkeley, Ca 94720	1
Mr. C. N. Bennett Naval Coastal Systems Lab Code P761 Panama City, FL 32401	1

	No. of Copies
Prof. L. N. Bhat Comput Sci/Oper Res Ctr Southern Methodist Univ. Dallas, TX 75275	1
Prof. W. R. Blischke Dept. of Quantitative Business Analysis Univ. of Southern California Los Angeles, CA 90007	1
J. E. Boyer, Jr. Dept. of Statistics Southern Methodist University Dallas, TX 75275	1
Dr. J. Chandra U.S. Army Research P.O. Box 12211 Research Triangle Park North Carolina 27706	1
Prof. H. Chernoff Dept. of Mathematics Mass. Inst. Tech. Cambridge, Mass 02139	1
Prof. C. Derman Dept. of Civil Engineering and Eng. Mech. Columbia University New York, New York 10027	1
Prof. R. L. Disney Virginia Polytechnic Inst. and State Univ. Dept. of Ind. Eng. Blacksburg, VA 24061	1
Mr. J. Dowling Defense Logistics Studies Information Exchange Army Logistics Management Center Fort Lee, VA 20390	1
Dr. M. J. Fischer Defense Communications Agency 1860 Wiehle Ave. Reston, VA 22070	1

	No. of Copies
Mr. Gene H. Gleissner Appl. Math. Lab. David Taylor Naval Ship Research and Development Ctr Bethesda, MD 20084	1
Prof. S. S. Gupta Dept. of Statistics Purdue University Lafayette, IN 47907	1
Prof. D. L. Hanson Dept. of Math. Sci. State University of New York, Binghamton, NY 13901	1
Prof. F. J. Harris Dept. of Elec. Eng. San Diego State University San Diego, CA 92182	1
Prof. L.H. Herbach Dept. of Oper. Res. and Sys. Anal. Polytechnic Inst. of N. Y. Brooklyn, NY 11201	1
Prof. M. J. Hinich Dept. of Economics Virginia Polytechnic Inst. and State Univ. Blacksburg, VA 24061	1
Prof. W. M. Hirsch Inst. of Math. Sci. New York Univ. New York, New York 10053	1
Prof. D. L. Iglehart Dept. of Oper. Res. Stanford Univ. Stanford, CA 94350	1
Prof. J. B. Kadane Dept. of Statistics Carnegie-Mellon Pittsburgh, PA 15213	1

	No. of Copies
Dr. Richard Lau, Director Office of Naval Research Branch Office 1030 East Green Street Pasadena, CA 91101	1
Prof. N. Leadbetter Dept. of Statistics University of North Carolina Chapel Hill, No. Carolina 27514	1
Prof. R. S. Levenworth Dept. of Ind. and Sys. Eng. University of Florida Gainsville, FL 32611	1
Prof. G. Lieberman Dept. of Oper. Res. Stanford University Stanford, CA 94305	1
Dr. N. R. Mann Biomathematics Dept. School of Public Health University of California Los Angeles, CA 90024	1
Dr. W. H. Marlow Program in Logistics George Washington University 707 22nd Street, N.W., Washington, D.C. 20037	1
Prof. J. A. Muckstadt Dept. of Oper. Res. Cornell University Ithaca, N.Y. 19850	1
Dr. Janet M. Myhre Ins. Decision Science for Business and Public Policy Claremont Men's College Claremont, CA 91711	1
B. S. Orleans Naval Sea Systems Command (Sea 03F), Rm 10SC8 Arlington, VA 20360	1
Prof. D. B. Owen Dept. of Statistics Southern Methodist University Dallas, TX 75222	1

	No. of Copies
Prof. E. Parzen Statistics Dept. Texas A.&M. College Station, TX 77843	1
Prof. S. L. Phoenix Sibley School of Mech. and Aerospace Eng. Cornell University Ithaca, NY 14850	1
Prof. M. L. Puri Dept. of Mathematics P.O. Box F Indiana University Foundation Bloomington, IN 47401	1
Prof. M. Rosenblatt Dept. of Mathematics University of California San Diego, CA 92093	1
Prof. S. M. Ross College of Engineering University of California Berkeley, CA 94720	1
Prof. L. L. Scharf Jr. Dept. of Elec. Eng. Colorado State University Ft. Collins, CO 89521	1
Prof. D. C. Siegmund Dept. of Statistics Stanford University Stanford, Ca. 94305	1
Prof. W. L. Smith Dept. of Statistics Univ. of North Carolina Chapel Hill, N.C. 27514	1
Prof. J. R. Thompson Dept. of Math. Sci. Rice University Houston, TX 77001	1
Prof. J. L. Tukey Dept. of Statistics Princeton University Princeton, N.J. 98540	1
Daniel H. Wagner Station Square One Paoli, PA 19301	1

	No. of Copies
Prof. Grace Wahba Dept. of Statistics University of Wisconsin Madison, WI 53706	1
Prof. G. S. Watson Dept. of Statistics Princeton University Princeton, N.J. 08540	1
Prof. Peter Bloomfield Dept. of Statistics Princeton University Princeton, N.J. 08540	1
Dr. D. R. Cox Dept. of Mathematics Imperial College London SW7 2BZ, ENGLAND	1
Prof. George S. Fishman Cur. in OR and Sys Analysis University of North Carolina Phillips Annex Chapel Hill, N.C. 20742	1
Dr. R. Gnanadesikan Bell Telephone Lab. Murray Hill, N.J.	1
Dr. A. J. Goldman, Chief OR Div. 205.02, Admin. A428 U.S. Dept. of Commerce Washington, D.C. 20234	1
Dr. H. Kobayashi IBM Yorktown Heights, NY 10598	1
Dr. John Lehoczky Dept. of Statistics Carnegie-Mellon Univ. Pittsburgh, PA 15213	1
Dr. A. Lemoine 1020 Guinda St. Palo Alto, CA 94301	1
Dr. J. MacQueen University of California Los Angeles CA 90024	1

	No. of Copies
Dr. M. Mazumdar Math. Dept. Westinghouse Labs. Churchill Boro Pittsburgh, PA 15235	1
Dr. Paul Schwitzer Graduate School of Management University of Rochester Rochester, NY 14627	1
Dr. Roy Welsch M.I.T., Sloan School Cambridge, MA 02139	1
Dr. Bruce Trumbo National Science Foundation Math. Sci. Div. 18th and "G", N.W. Washington, D.C. 20550	1
Dr. William Brogan National Science Foundation Math. Sci. Div. 18th and "G", N.W. Washington, D.C. 20550	1
H. E. Bamford, Jr., Program for Research on Information Systems National Science Foundation 18th and "G", N.W. Washington, D.C. 20550	1
Prof. Marcel F. Neuts University of Delaware Department of Mathematics Newark, Delaware 19711	1
Mr. John J. Bruggeman 340 Vallejo Drive No. 58 Millbrae, CA 94030	1
Gregory Prastacos Dept. of Decision Sciences The Wharton School/CC University of Pennsylvania Philadelphia, PA 19104	1

	No. of Copies
Professor John S. Rose Dept. of Management Systems University of Richmond, Richmond, VA 23173	1
Professor Stephen C. Graves Alfred P. Sloan School of Management Massachusetts Institute of Technology 50 Memorial Drive, Cambridge, MA 02139	1
Yoshio Mizoroki Dept. of Management Sciences Illinois Institute of Technology 3300 S. Federal Chicago, ILL 60616	1
Prof. Henk Tijjns Mathematisch Centrum Dept. of Operations Research 2e Boerhaavestraat 49, Amsterdam The Netherlands	1
Library, Code 55 Naval Postgraduate School Monterey, CA 93940	1
Naval Postgraduate School Monterey, CA 93940	
AttN: Professor J. D. Esary, Code 55Ey	1
Professor P.A.W. Lewis, Code 55Lw	1
Professor M.G. Sovereign, Code 55Zo	1
Professor D. P. Gaver, Code 55Gv	5
Professor P. A. Jacobs, Code 55Jc	10
R. J. Stampfel, Code 55	1

U-187,376
Naval Postgraduate School
NPS55-79-07.
Storage problems when
nothing", by D. P. Garver
March 1979. 32 p.

U187376

DUDLEY KNOX LIBRARY - RESEARCH REPORTS



5 6853 01062194 9